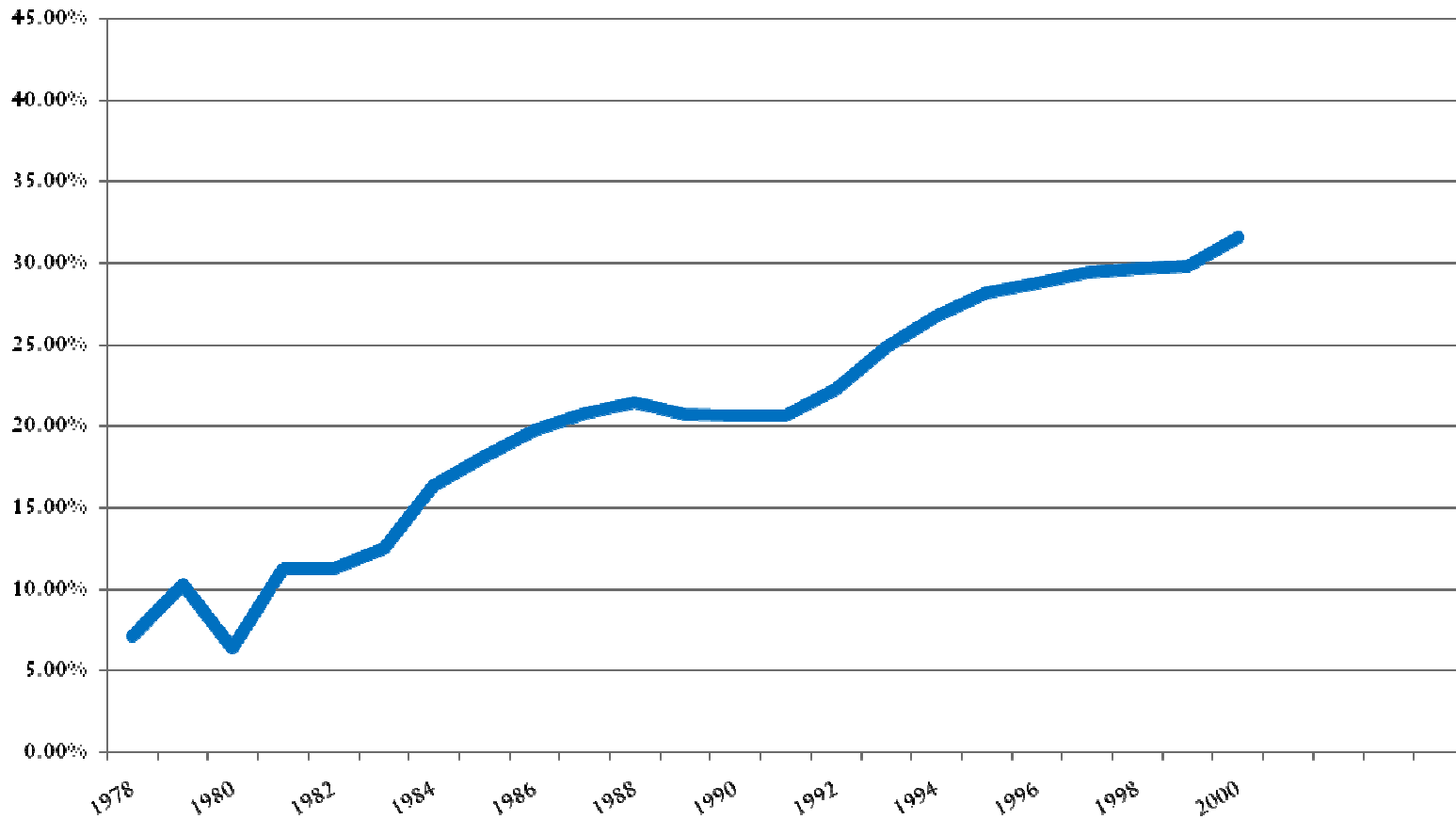


Off-Farm Employment and Educational Attainment in Rural China

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Background

- Increased Off-Farm Employment



Source: China SSB

Background

- Rising Internal Labor Migration
 - 1995: 54 Million (Rozelle et al., 1999)
 - 1996: 68 Million (Liang, 2001)
 - 2005: >100 Million (Omelaniuk, 2005)
- Fastest Growing Component of Off-Farm Employment (de Brauw et al., 2002)

Background

- Higher Enrollment/Literacy Rates in Rural Areas
 - Persistent Rural/Urban Education Gap (Connelly and Zheng, 2003; Zhang and Kanbur, 2003)
- Possible link to increased incidence of migration (De Brauw and Giles, 2006)

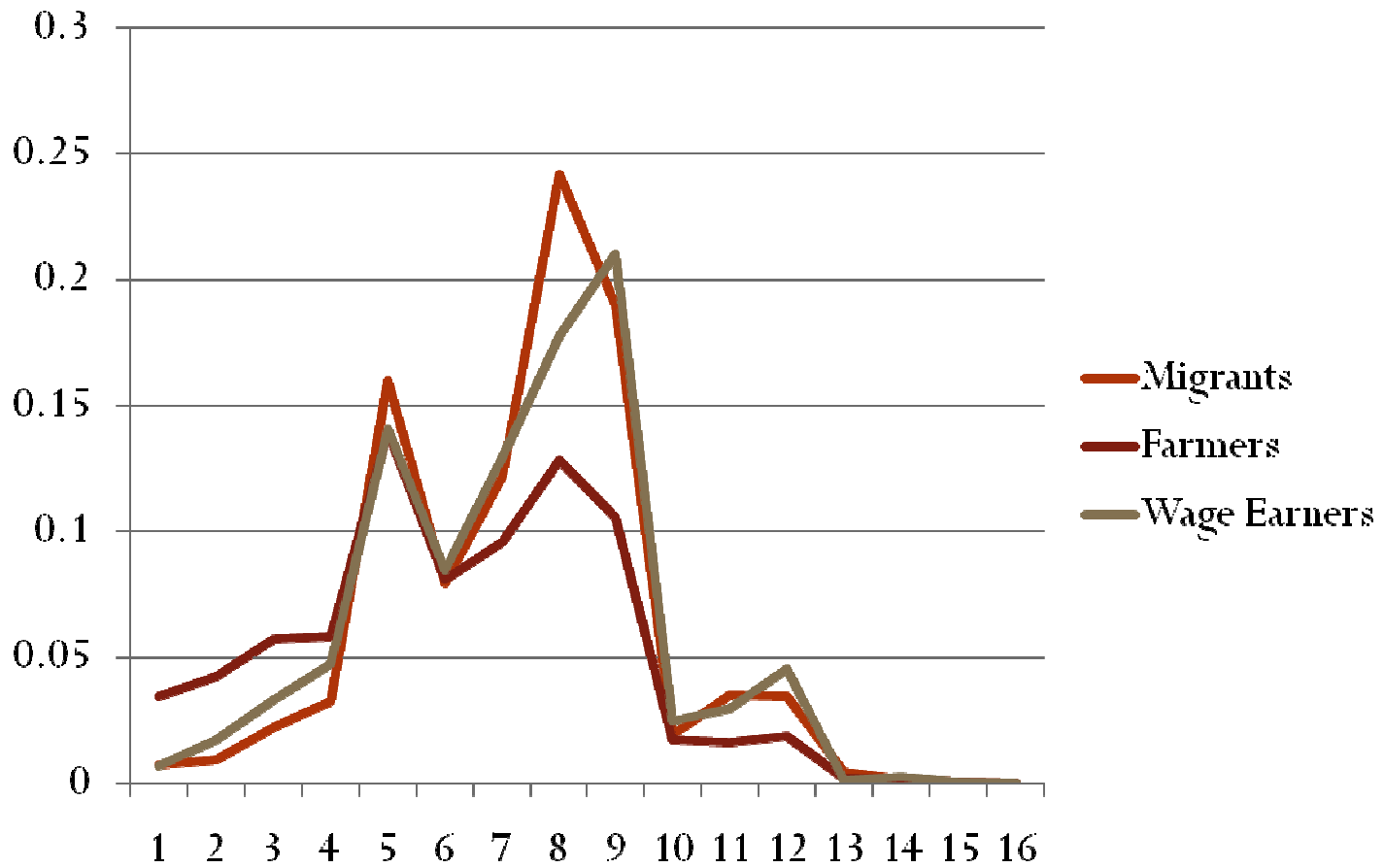
Questions

- How is human capital rewarded in the each of three rural sectors: Migration, TVE Employment, and Household Farming?
- How has increased access to off-farm employment opportunities affected educational attainment in rural areas?

Data

- China Household Income Project (CHIP), 1995 Rural Sample.
 - 7,998 Households, 34,739 Individuals
 - 19 Provinces, 112 Counties

Preliminary Evidence



Preliminary Evidence

Province Level Correlation

	<u>Average Years of Schooling</u>	<u>College/Professional Students</u>	<u>Coll/Prof/Technical Students</u>
Migrant % LLF	-0.37	-0.455	-0.40
LL % LLF	0.81	0.52	0.80

County Level Correlation

	<u>Average Years of Schooling</u>	<u>College/Professional Students</u>	<u>Coll/Prof/Technical Students</u>
Migrant % LLF	-0.12	-0.15	-0.17
LL % LLF	0.42	0.05	0.36

Household Time Allocation

- Household heads are uncertain as to how a child's labor time will be allocated in the future

$$E[w_{c,t}] = E[T_{f,t}] \cdot w_t^F(H_t \alpha_F, \mathbf{Z}_F) + E[T_{l,t}] \cdot w_t^l(H_t \alpha_l, \mathbf{Z}_l) + E[T_{M,t}] \cdot w_t^M(H_t \alpha_M, \mathbf{Z}_M)$$

Household Time Allocation

- Positive returns to human capital in a given sector are not sufficient to promote higher levels of educational attainment
- Returns must be high enough to offset the opportunity cost of time spent in education

Analysis

- The empirical analysis consists of two sections:
 1. Derive the returns to education in Household Farming, Migrant Labor, TVE Employment
 2. Determine how potential employment in each sector affects educational attainment.

Analysis - Part One

- Farm income is estimated using a simple value-added function.
- Income from migrant labor and local wage earning will be modeled using Mincer-type income equations.

Farm Income

- Farm income is estimated using a two-stage value added equation:

1)
$$\ln(V) = \alpha + \psi_1 FarmDays + \psi_2 AvgSchYrs + \psi_3 Land + \psi_4 FlatLand + \psi_5 K_F + \psi_6 PctLandIrr + \zeta_1 Impov + \varepsilon,$$

2)
$$FarmDays = \alpha + \phi_1 AvgSchYrs + \phi_2 Land + \phi_3 FlatLand + \phi_4 K_F + \phi_5 Workers + \phi_6 PartyinHH + \phi_7 PctLandIrr + \zeta_1 Impov + \zeta_3 MigPctLF + \zeta_4 LWEPctLF + \varepsilon.$$

Farm Income

- Days allocated to farming may be endogenously determined.
- Time allocated to farming is instrumented using:
 - Total household labor force (*HHWorkers*)
 - Proportion of the local labor force participating in the local wage earning market (*LLPctLF*)
 - Proportion of the local labor force reporting migrant labor force participation (*MigrantPctLF*)

Results

	<u>DV = ln(V)</u>	<u>DV = FarmDays</u>
<i>FarmDays</i>	0.0009 (16.33)**	
<i>AvgSchYrs</i>	0.005 (2.10)*	-0.54 (-0.64)
<i>Land</i>	0.03 (22.42)**	3.22 (8.03)**
<i>LandPctIrr</i>	0.08 (5.48)**	16.49 (3.64)**
<i>FlatLand</i>	0.16 (11.26)**	17.83 (3.83)**
K_F	0.003 (9.40)**	0.45 (4.43)**
<i>Impov</i>	-0.20 (-11.63)**	43.04 (7.91)**
<i>Workers</i>		95.22 (53.18)**
<i>PartyinHH</i>		-11.93 (-2.23)*
<i>MigPctLF</i>		.05 (0.14)
<i>LWEPctLF</i>		-3.74 (-18.18)**
<u><i>Adj r²</i></u>	0.27	0.41

Farm Income

- Main Findings:
 - Low return to average household level of education (0.5%)

- Farming Seems to be an Occupation of Last Resort
 - Greater local off-farm employment opportunities reduce time allocated to farming.
 - More time is allocated to farming in impoverished counties

Local Wage Earning

- Income for local wage earning and migrant labor will be estimated using a Mincer-type income equation
- The Heckman two-step method is used to control for selection bias
 - Step 1: Estimate a probit for local wage participation.
 - Step 2: Estimate income equation using the Inverse Mill's Ratio (IMR) calculated from the probit results
 - Tobit estimation results are included to show how each variable used in the probit function affects variation in time allocation rather than likelihood of participation

Local Wage Earning

- Estimated income equation:

1.
$$Z = \alpha + \beta_1 SchYrs + \beta_2 Female + \gamma_1 PartyinHH + \gamma_2 LandPerCap + \gamma_3 FlatLand + \gamma_4 K_F PerCap + \delta_1 Impov + \delta_2 MigPctLF + \delta_3 LWEPctLF + \varepsilon.$$

2.
$$\ln(DW) = \alpha + \tau_1 SchYrs + \tau_2 Exp + \tau_3 Exp^2 + \tau_4 Female + \omega_1 \lambda + \varepsilon.$$

Results

	Local Wage Earning		
	DV=ln(DW)	Probit	Tobit
<i>SchYrs</i>	0.02 (2.05)*	0.001 (3.88)**	9.63 (4.15)**
<i>Exp</i>	0.06 (4.73)**		
<i>Exp2</i>	-0.002 (-4.13)**		
<i>Female</i>	-0.05 (0.69)	-0.03 (-13.62)**	-196.55 (-12.69)**
<i>PartyinHH</i>	-0.19 (-2.71)**	0.009 (3.90)**	63.04 (3.93)
<i>LandPerCap</i>		-0.002 (-2.74)**	-22.24 (-3.49)**
<i>FlatLand</i>		0.003 (1.76) ⁺	34.05 (2.05)*
<i>K_FPerCap</i>		-0.000008 (-3.13) ⁺	-0.06 (-3.23)**
<i>Impov</i>		0.001 (0.41)	-4.88 (-0.21)
<i>MigPctLF</i>		0.0003 (1.80) ⁺	1.87 (1.31)
<i>LWEPctLF</i>		0.002 (21.95)**	12.05 (19.73)**
<i>IMR</i>	-0.41 (-5.39)**		
		P > chi ² = 0	Pseudo R ² = .10

Local Wage Earning

- Main Findings:
 - Positive returns to education:
 - Income (2%)
 - Participation
 - Females are less likely to participate
 - Evidence of selection bias in the sample

Migrant Labor

- Migrant earnings are not directly observed in the data set
 - Estimate migrant contribution to household income via remittance instead
- Imperfect substitute for migrant earnings
 - Must assume that individuals of equal education levels remit equal proportions of their income

Migrant Labor

- Migrant earnings are modeled using an income equation similar to the local wage earning model:

1. $DR = \alpha + \tau_1 SchYrs + \tau_2 Exp + \tau_3 Exp^2 + \tau_4 Female + \omega_1 \lambda + \varepsilon.$

2. $Z = \alpha + \beta_1 SchYrs + \beta_2 Female + \gamma_1 PartyinHH + \gamma_2 LandPerCap + \gamma_3 FlatLand + \gamma_4 K_F PerCap + \delta_1 Impov + \delta_2 MigPctLF + \delta_3 LWEPctLF + \varepsilon.$

Results

	Migrant Labor		
	DV = DR	Probit	Tobit
<i>SchYrs</i>	-0.20 (-2.36)*	0.01 (16.09)**	25.03 (16.33)**
<i>Exp</i>	-0.07 (-0.51)		
<i>Exp2</i>	0.007 (0.98)		
<i>Female</i>	-0.55 (-1.05)	-0.06 (-17.50)**	-145.88 (-15.86)**
<i>PartyinHH</i>		-0.006 (-1.33)	-4.58 (-0.41)
<i>LandPerCap</i>		-0.003 (-2.00)*	-8.22 (-2.19)**
<i>FlatLand</i>		-0.008 (1.95) ⁺	-19.94 (-1.96)*
<i>K_fPerCap</i>		-0.000004 (-1.15)	-0.01 (1.09)
<i>Impov</i>		0.0007 (0.16)	-1.07 (-0.10)
<i>MigPctLF</i>		0.006 (21.90)**	15.66 (21.05)**
<i>LWEPctLF</i>		-0.001 (-3.00)**	-1.35 (-2.60)**
<i>IMR</i>	-3.63 (4.75)**		
		P > chi ² = 0	Pseudo R ² = .05

Migrant Labor

- Main Findings:
 - Migrants with higher education levels remit less income, but higher education levels increase the probability of migration
 - Females are less likely to migrate
 - Individuals reporting more land per household member spend less time migrating
 - Larger local migrant networks increase the probability of migration while a larger local wage sector discourages migration

Overall Results

- Both migration and TVE employment were positively related to education
- Migrant labor showed negative returns to household income via remittance, but the effect on migrant income is unclear
- The returns to education are higher in TVE employment than in household farming

Education

- The results indicate positive returns to education in both off-farm sectors, but we still do not know if they are high enough to encourage investment in education.
- To determine whether or not these returns are sufficient to encourage investment in education, we model educational attainment directly

Education

- To determine the effect of different employment opportunities on educational attainment, we use a Cox proportional hazards model
- $h(t)$ represents the hazard of any individual i dropping out of school at time t :

$$h_i(t) = h_0(t) \exp(\beta \mathbf{V})$$

Results

	(1)	(2)
	Years 1-12	Years 5-10
<i>Impov</i>	0.977 (-0.31)	0.983 (-0.20)
<i>NetHHInc</i>	1.000 (-0.92)	1.00 (-0.13)
<i>AvgPrntSch</i>	0.906 (-8.87)**	0.915 (-7.30) ⁺
<i>K_FPerCapita</i>	1.000 (2.18)*	1.000 (1.54)
<i>LandPerCapita</i>	1.006 (0.25)	1.035 (1.44)
<i>LandPcntIrr</i>	0.980 (-1.29)	0.988 (-0.67)
<i>FlatLand</i>	1.07 (1.11)	1.136 (1.91) ⁺
<i>MigPctLF</i>	1.018 (3.55)**	1.024 (4.58)**
<i>PartyinHH</i>	0.840 (-2.51)*	0.87 (-1.81) ⁺
<i>LWEPctLF</i>	0.997 (-0.87)	0.998 (-0.50)
<i>Female</i>	1.32 (5.42)**	1.287 (4.56)**
	P>chi ² = 0.00	P>chi ² = 0.00

Education

- Main Findings:
 - A larger local migrant labor force increases the risk of dropping out of school
 - Females are more likely to drop out of school
 - Higher levels of parental schooling reduce the risk of dropping out
 - Party membership reduces the risk of dropping out

Conclusions

- Migrant labor rewards human capital but not as well as local wage employment.
- Migration may help raise household incomes in the short-run, but promoting the development of rural industries may be more beneficial for long-term growth
 - Human capital externalities

Future Work

- Experiment with different measurements of human capital
- Incorporate risk as an element of household income maximization
- Use panel data set to relax assumptions forced on the proportional hazards model